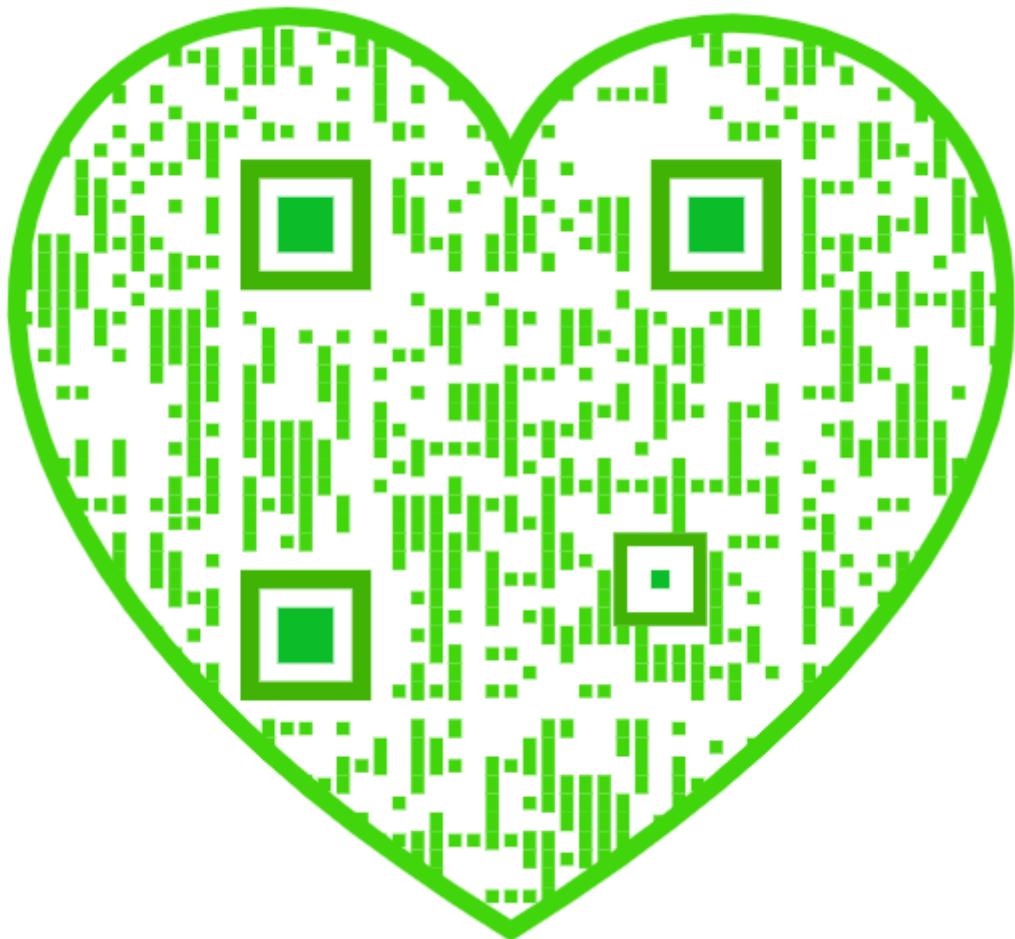


Master in Artificial Intelligence



Algorithm Selection & Development II



Purpose

The purpose of the section is to help you learn how to research, select, and develop appropriate algorithms to become a Successful Artificial Intelligence (AI) Engineer

At the end of this lecture, you will learn the following

- How to research, select, and develop appropriate machine learning algorithms or deep learning architectures based on the problem at hand and the available data?**



How to determine type of output and evaluation metrics?

Understand the problem

Define the problem as a

Determine

Domain

Supervised

type of output
(e.g., classification,
regression,
clustering)

Objectives

Unsupervised

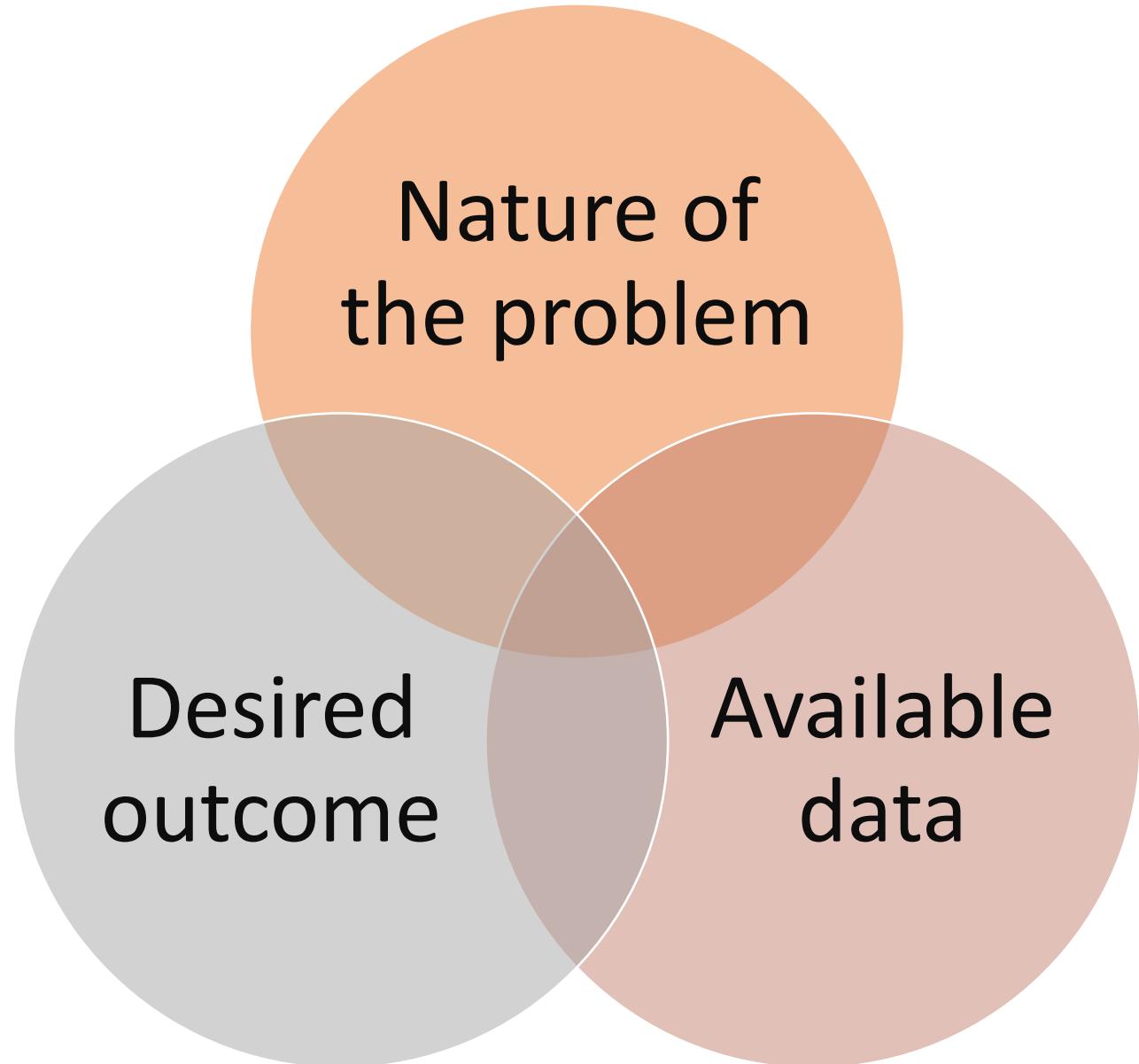
Evaluation metrics

Constraints

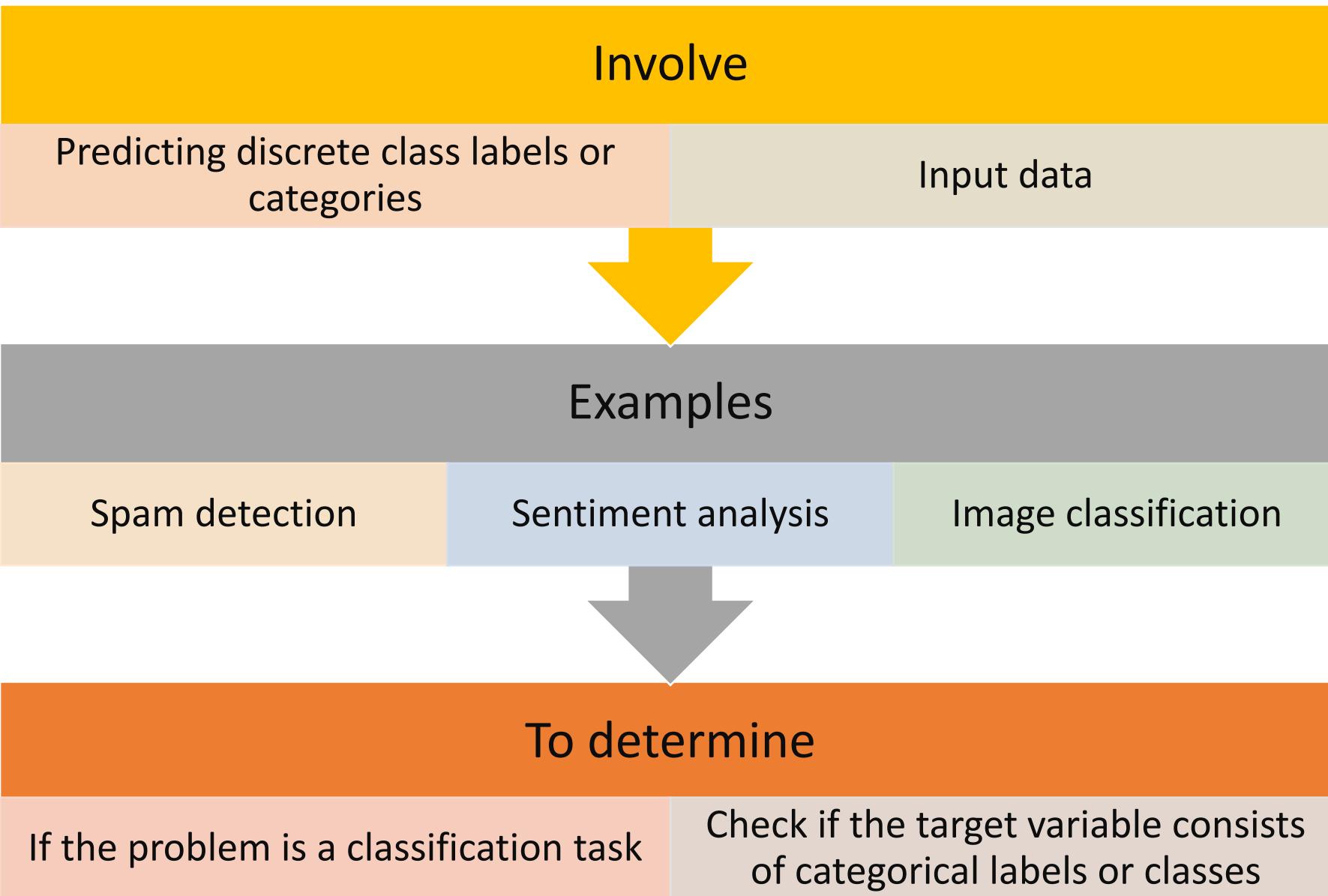
Reinforcement learning task



How to determine type of output and evaluation metrics?



Classification



Classification- Evaluation Metrics

Evaluation Metrics

Accuracy

Precision, Recall, and F1-score

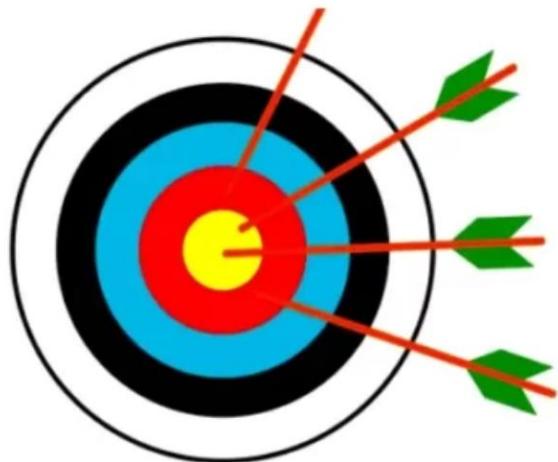
Area Under the Receiver Operating Characteristic (ROC-AUC) curve

Confusion Matrix

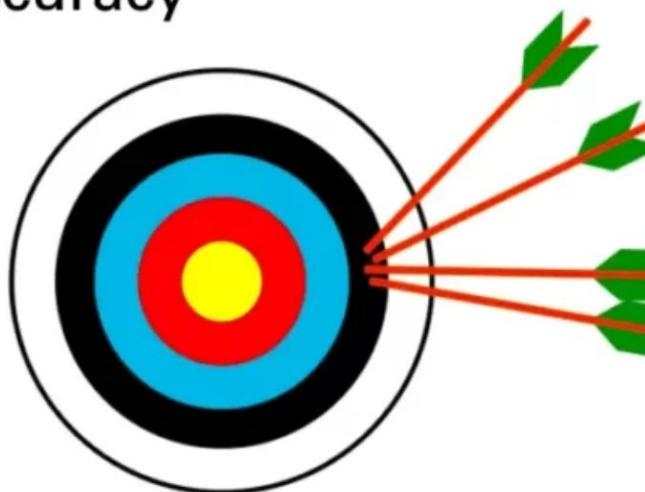


Accuracy and Precision

Precision VS Accuracy



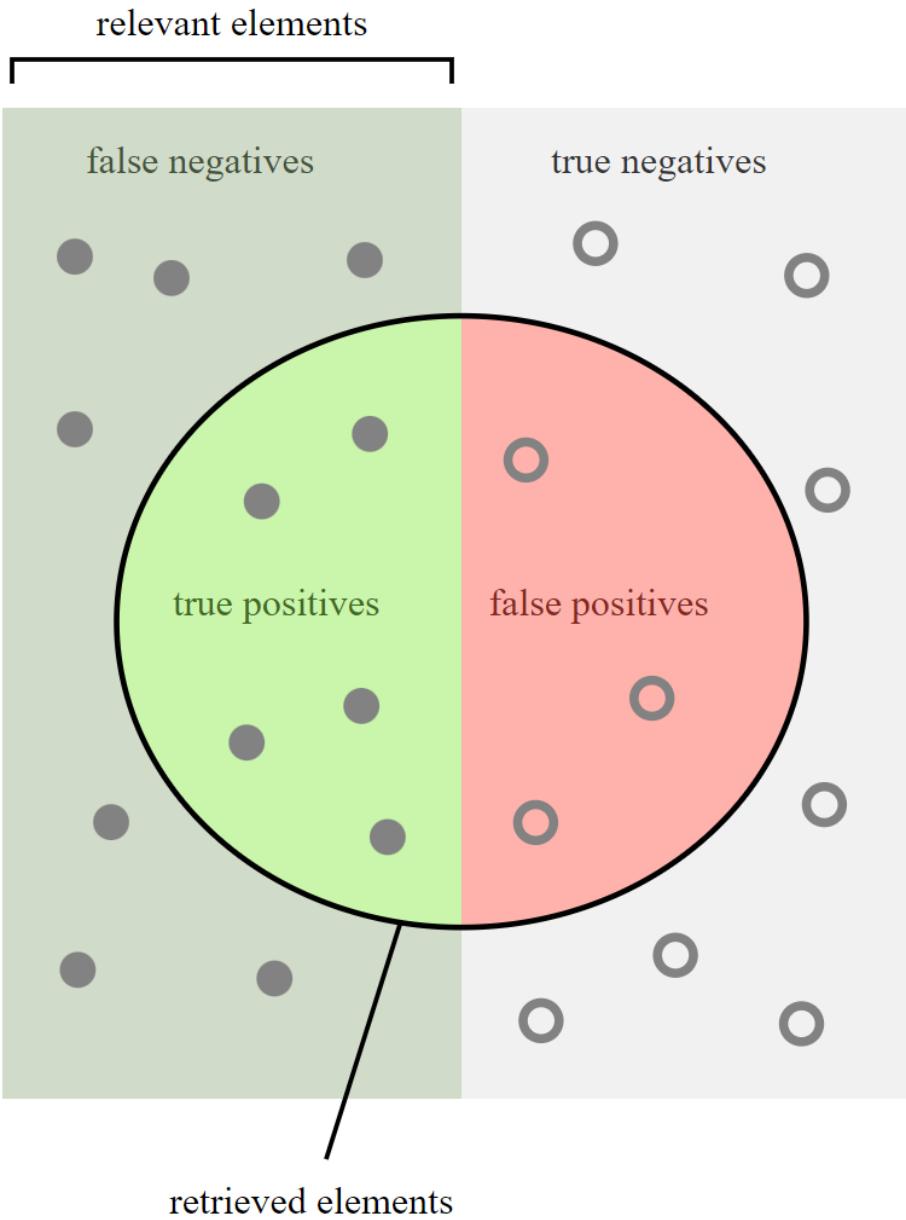
High Accuracy



High Precision

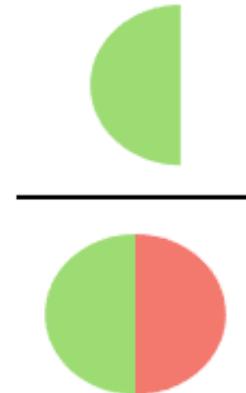


Recall



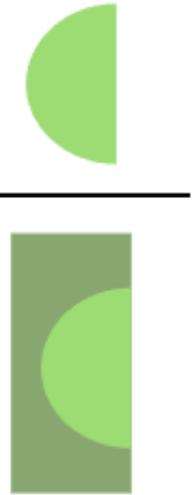
How many retrieved items are relevant?

Precision = $\frac{\text{true positives}}{\text{true positives} + \text{false positives}}$



How many relevant items are retrieved?

Recall = $\frac{\text{true positives}}{\text{true positives} + \text{false negatives}}$



F1 Score

F1 score Formula

$$F1 \text{ score} = \frac{2}{\frac{1}{Precision} + \frac{1}{Recall}} = 2 \cdot \frac{Precision * Recall}{Precision + Recall}$$

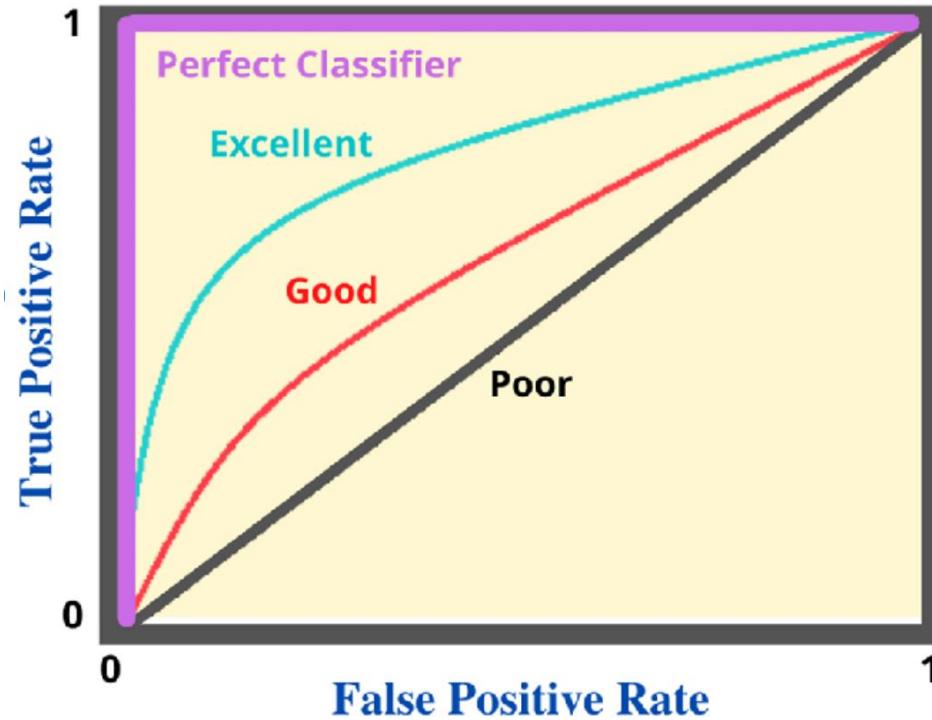
$$\Rightarrow F1 \text{ score} = 2 \cdot \frac{Precision * Recall}{Precision + Recall}$$

1. The value of the F1 score ranges from 0 to 1.

1. 0: Worst Case
2. 1: Best Case



Area Under the Receiver Operating Characteristic (ROC-AUC) curve



		True Class	
		T	F
Acquired Class	Y	True Positives (TP)	False Positives (FP)
	N	False Negatives (FN)	True Negatives (TN)

$$\text{True Positive Rate (TPR)} = \frac{TP}{TP + FN}$$

$$\text{False Positive Rate (FPR)} = \frac{FP}{FP + TN}$$

Accuracy

$$(ACC) = \frac{TP + TN}{TP + FP + TN + FN}$$



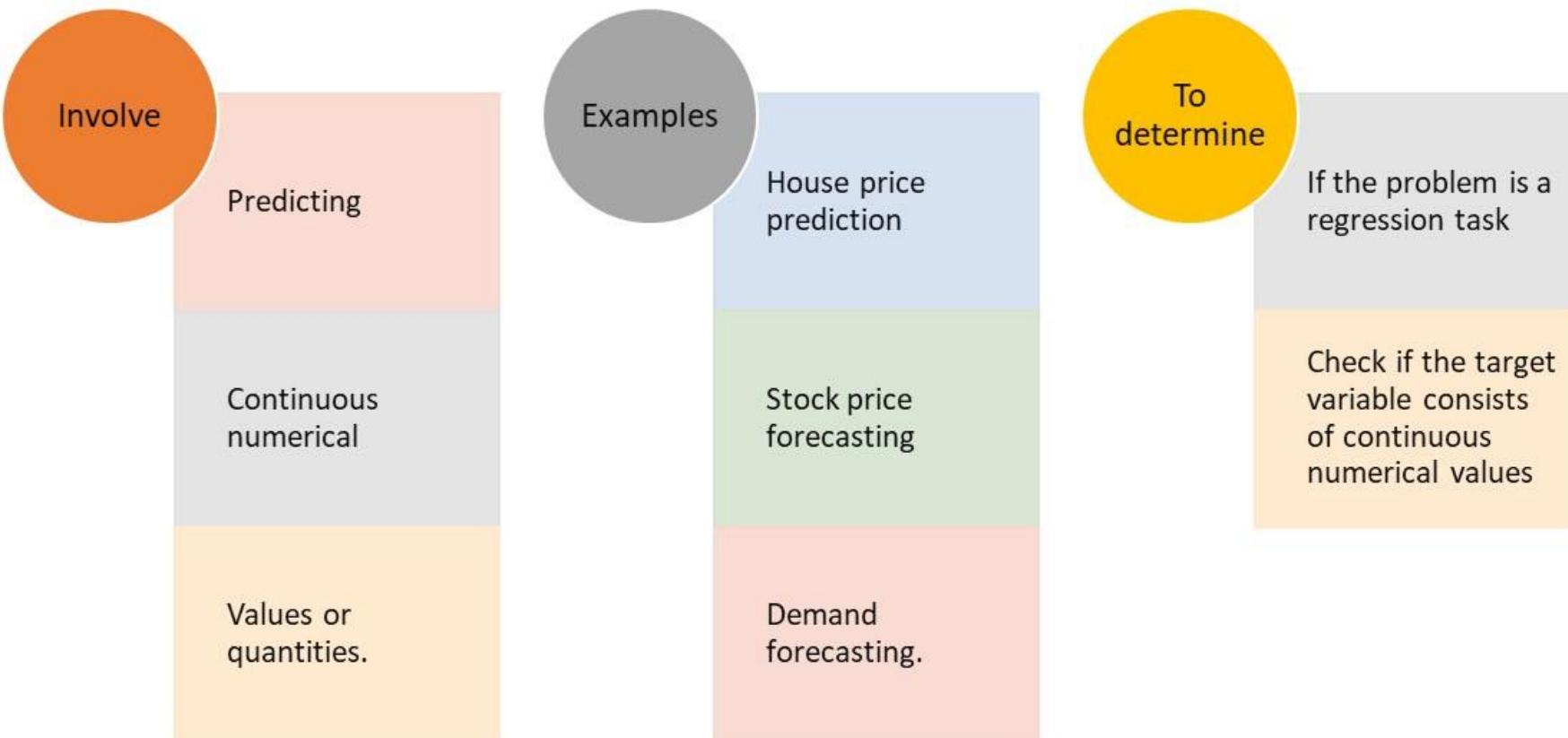
Confusion Matrix

		Positive	Negative
Positive	Negative	TP	FN (Type II Error)
	Positive	FP (Type I Error)	TN
RECALL		SENSITIVITY	
TPR = $\frac{TP}{TP + FN}$	FALSE POSITIVE	FNR = $\frac{FN}{TP + FN}$	TRUE NEGATIVE
f1-score (y-test, y-pred)		Accuracy = $\frac{TP + TN}{TP + FN + FP + TN}$	
$fPR = \frac{FP}{TN + FP}$	TRUE POSITIVE	TNR = $\frac{TN}{TN + FP}$	from sklearn.metrics import f1_score
RECALL		FALSE NEGATIVE	



What is next?

Regression



Master in Artificial Intelligence



Algorithm Selection & Development II